

Last Time: Orthogonality.

Gram-Schmidt Process:

Given lin. ind. vech. v_1, v_2, \dots, v_k in \mathbb{R}^n , we can construct a set of mutually orthogonal vech. u_1, u_2, \dots, u_k with the same span. Formulaically:

$$\begin{cases} u_1 = v_1 \\ u_i = v_i - \text{proj}_{u_1}(v_i) - \text{proj}_{u_2}(v_i) - \dots - \text{proj}_{u_{i-1}}(v_i) \end{cases}$$

Ex: Apply GS-process to $v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$, $v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$, $v_3 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}$.

Sol: $u_1 = v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$.

$$\begin{aligned} u_2 &= v_2 - \text{proj}_{u_1}(v_2) = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{0}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} u_3 &= v_3 - \text{proj}_{u_2}(v_3) - \text{proj}_{u_1}(v_3) \\ &= \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{-1}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{4}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 & +\frac{1}{2} & -\frac{3}{2} \\ 3 & +0 & -\frac{3}{2} \\ 1 & -\frac{1}{2} & -\frac{5}{2} \end{pmatrix} = \begin{pmatrix} -5/6 \\ 5/3 \\ -5/6 \end{pmatrix} = \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \end{aligned}$$

Check: $u_1 \cdot u_2 = 0$, $u_1 \cdot u_3 = 0$, $u_2 \cdot u_3 = 0$

$$u_1 \cdot u_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = 1 + 0 - 1 = 0$$

$$u_1 \cdot u_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} = \frac{5}{6} (-1 + 2 - 1) = \frac{5}{6} \cdot 0 = 0$$

$$u_2 \cdot u_3 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} = \frac{5}{6} (-1 + 0 + 1) = \frac{5}{6} \cdot 0 = 0 \quad \checkmark \quad \square$$

Another check method: Note $u \cdot v = u^T v$

$$\text{(i.e. } \begin{pmatrix} x \\ y \\ z \end{pmatrix} \cdot \begin{pmatrix} a \\ b \\ c \end{pmatrix} = (x \ y \ z) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = [xa + yb + zc])$$

Take $A = [u_1 | u_2 | u_3]$, check

$$A^T A = \begin{bmatrix} u_1^T \\ u_2^T \\ u_3^T \end{bmatrix} [u_1 | u_2 | u_3] = \begin{bmatrix} u_1^T u_1 & u_1^T u_2 & u_1^T u_3 \\ u_2^T u_1 & u_2^T u_2 & u_2^T u_3 \\ u_3^T u_1 & u_3^T u_2 & u_3^T u_3 \end{bmatrix}$$


$$= \begin{bmatrix} u_1 \cdot u_1 & u_1 \cdot u_2 & u_1 \cdot u_3 \\ u_2 \cdot u_1 & u_2 \cdot u_2 & u_2 \cdot u_3 \\ u_3 \cdot u_1 & u_3 \cdot u_2 & u_3 \cdot u_3 \end{bmatrix} = \begin{bmatrix} |u_1|^2 & 0 & 0 \\ 0 & |u_2|^2 & 0 \\ 0 & 0 & |u_3|^2 \end{bmatrix}$$

if the u_i 's are mutually orthogonal...

Point: $A^T A$ is a diagonal matrix if columns of A are mutually orthogonal... Should do: normalize the columns

of A (i.e. force $|u_i| = 1$ for all i by taking suitable scalar multiples), then we obtain an "orthogonal matrix".

Defn: A matrix M is orthogonal when $M^T = M^{-1}$ (M is $n \times n$).

Prop: M is orthogonal if and only if the columns of M form an orthonormal basis for \mathbb{R}^n . pf: Easy exercise 

Defn: A basis of \mathbb{R}^n is orthonormal when the elements are mutually orthogonal and all have length 1.

Ex: Moment ago: we computed $u_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$, $u_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$, $u_3 = \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix}$ form an orthogonal basis of \mathbb{R}^3 . However,

$$|u_1| = \sqrt{1^2 + 1^2 + 1^2} = \sqrt{3}, \quad |u_2| = \sqrt{1+1} = \sqrt{2},$$

$$|u_3| = \frac{5}{6} \sqrt{1+4+1} = \frac{5}{6} \sqrt{6} = \frac{5}{\sqrt{6}}, \text{ so } B = \{u_1, u_2, u_3\} \text{ is}$$

not orthonormal... But

$$\begin{aligned} \hat{B} &= \left\{ \frac{1}{|u_1|} u_1, \frac{1}{|u_2|} u_2, \frac{1}{|u_3|} u_3 \right\} = \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{5}{5\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\} \\ &= \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\} \stackrel{IS}{=} \text{orthonormal!} \\ &= \left\{ \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}, \begin{pmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{pmatrix}, \begin{pmatrix} -\frac{1}{\sqrt{6}} \\ \frac{2}{\sqrt{6}} \\ -\frac{1}{\sqrt{6}} \end{pmatrix} \right\} \end{aligned}$$

Check: $\begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix}^T \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \stackrel{???}{=} \underline{I}_3$

$$\begin{aligned} &\begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{6}} & \frac{2}{\sqrt{6}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{3} + \frac{1}{3} + \frac{1}{3} & \frac{1}{\sqrt{2}\sqrt{3}} + 0 - \frac{1}{\sqrt{2}\sqrt{3}} & \left[-\frac{1}{\sqrt{3}\sqrt{6}} + \frac{2}{\sqrt{3}\sqrt{6}} - \frac{1}{\sqrt{3}\sqrt{6}} \right] \\ \frac{1}{\sqrt{2}\sqrt{3}} + 0 - \frac{1}{\sqrt{2}\sqrt{3}} & \frac{1}{2} + 0 + \frac{1}{2} & \frac{1}{\sqrt{2}\sqrt{6}} + 0 + \frac{1}{\sqrt{2}\sqrt{6}} \\ \left[-\frac{1}{\sqrt{3}\sqrt{6}} + \frac{2}{\sqrt{3}\sqrt{6}} - \frac{1}{\sqrt{3}\sqrt{6}} \right] & \frac{1}{\sqrt{2}\sqrt{6}} + 0 + \frac{1}{\sqrt{2}\sqrt{6}} & \frac{1}{6} + \frac{4}{6} + \frac{1}{6} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \checkmark \quad \text{!!} \end{aligned}$$

Remark: This will always compute an orthonormal basis from orthogonal one.

Algorithm (Extended Gram-Schmidt Process): Given v_1, v_2, \dots, v_k l.i.n. indep. in \mathbb{R}^n . To compute an orthonormal collection w/ same span:

- ① Apply the Gram-Schmidt Process to v_1, v_2, \dots, v_k .
- ② Normalize each output vector (i.e. scale each u_i by $\frac{1}{|u_i|}$).

Ex: Apply extended GS process to $V_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$, $V_2 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}$, $V_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$.
 (NB: compare w/ previous example to note order matters for GS-process!)

Sol: $u_1 = V_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$

$$u_2 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \text{proj}_{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{-1}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 + \frac{1}{2} \\ 3 + 0 \\ 1 - \frac{1}{2} \end{pmatrix} = \begin{pmatrix} \frac{1}{2} \\ 3 \\ \frac{1}{2} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}$$

$$u_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \text{proj}_{\frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \text{proj}_{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{8}{38} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} - \frac{0}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{4}{19} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} = \frac{1}{19} \begin{pmatrix} 19-4 \\ 19-24 \\ 19-4 \end{pmatrix} = \frac{1}{19} \begin{pmatrix} 15 \\ -5 \\ 15 \end{pmatrix} = \frac{5}{19} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix}$$

$\text{proj}_{c\vec{u}}(\vec{v}) = \text{proj}_{\vec{u}}(\vec{v})$
 Span is equal!

GS Process yields $B = \left\{ \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}, \frac{5}{19} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix} \right\}$. Normalizing,

$$\hat{B} = \left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{38}} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{19}} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix} \right\}$$

$|c\vec{u}| = |c| |\vec{u}|$ so normalizing $c\vec{u} \mapsto \frac{1}{|c|} c\vec{u} = \frac{c}{|c| |\vec{u}|} \vec{u} = \frac{1}{|\vec{u}|} \vec{u}$
 when $c > 0$

Q: Why are we computing orthonormal bases?

A: Orthonormal collections give vectors very nice representations...

In the GS process: $u_i = V_i - \sum \text{proj}_{u_j}(V_i)$

$\leadsto V_i = \sum c_j u_j$

when u_j 's orthonormal

\star So constant $c_j u_j = \text{proj}_{u_j}(V_i) = \frac{u_j \cdot V_i}{u_j \cdot u_j} u_j = (u_j \cdot V_i) u_j$

Ex: The standard basis is an orthonormal basis.

$$v = \begin{pmatrix} a \\ b \\ c \end{pmatrix} = a e_1 + b e_2 + c e_3 = \underbrace{(v \cdot e_1)} e_1 + \underbrace{(v \cdot e_2)} e_2 + \underbrace{(v \cdot e_3)} e_3$$

Point: Orthonormal bases generalize the standard basis.

Ex: Compute $\text{Rep}_{\hat{B}} \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix}$ where $\hat{B} = \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\}$.

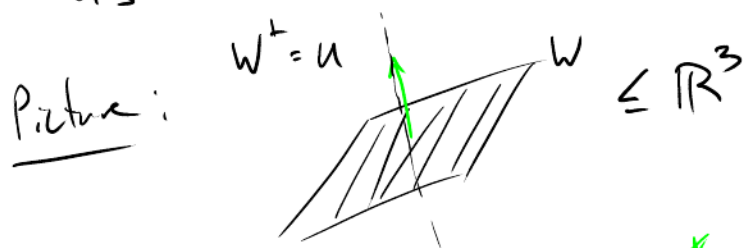
Sol: $v = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}$ has

$$\left. \begin{aligned} u_1 \cdot v &= \frac{1}{\sqrt{3}} (2 + 1 + 2) = \frac{5}{\sqrt{3}} \\ u_2 \cdot v &= \frac{1}{\sqrt{2}} (2 + 0 - 2) = 0 \\ u_3 \cdot v &= \frac{1}{\sqrt{6}} (-2 + 2 - 2) = -\frac{2}{\sqrt{6}} \end{aligned} \right\} \rightarrow \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix} = \frac{5}{\sqrt{3}} u_1 + 0 u_2 - \frac{2}{\sqrt{6}} u_3.$$



ORTHOGONAL COMPLEMENTATION

Defn: A complement of subspace $W \subseteq V$ is a subspace U such that every vector of V can be expressed uniquely as $v = w + u$ where $w \in W$ and $u \in U$.



Prop: If $W \subseteq \mathbb{R}^n$, then $W^\perp = \{ u \in \mathbb{R}^n : u \cdot w = 0 \text{ for all } w \in W \}$ is the complement of W .

Proof: Every basis of W extends to a basis of \mathbb{R}^n .

Pick B a basis of W . Apply Extended GS to obtain \hat{B} . \hat{B} is still a basis of W . Extend to

$A = \hat{B} \cup \hat{D}$ a basis for \mathbb{R}^n . $\hat{A} = \hat{B} \cup \hat{D}$. $W = \text{span}(\hat{B})$ and $W^\perp = \text{span}(\hat{D})$.

Computationally: to compute W^\perp :

① express $W = \text{Col}(A)$ for matrix A .

② $W^\perp = \text{null}(A^T)$

Point \rightarrow Use $A =$ matrix of any basis \smile

